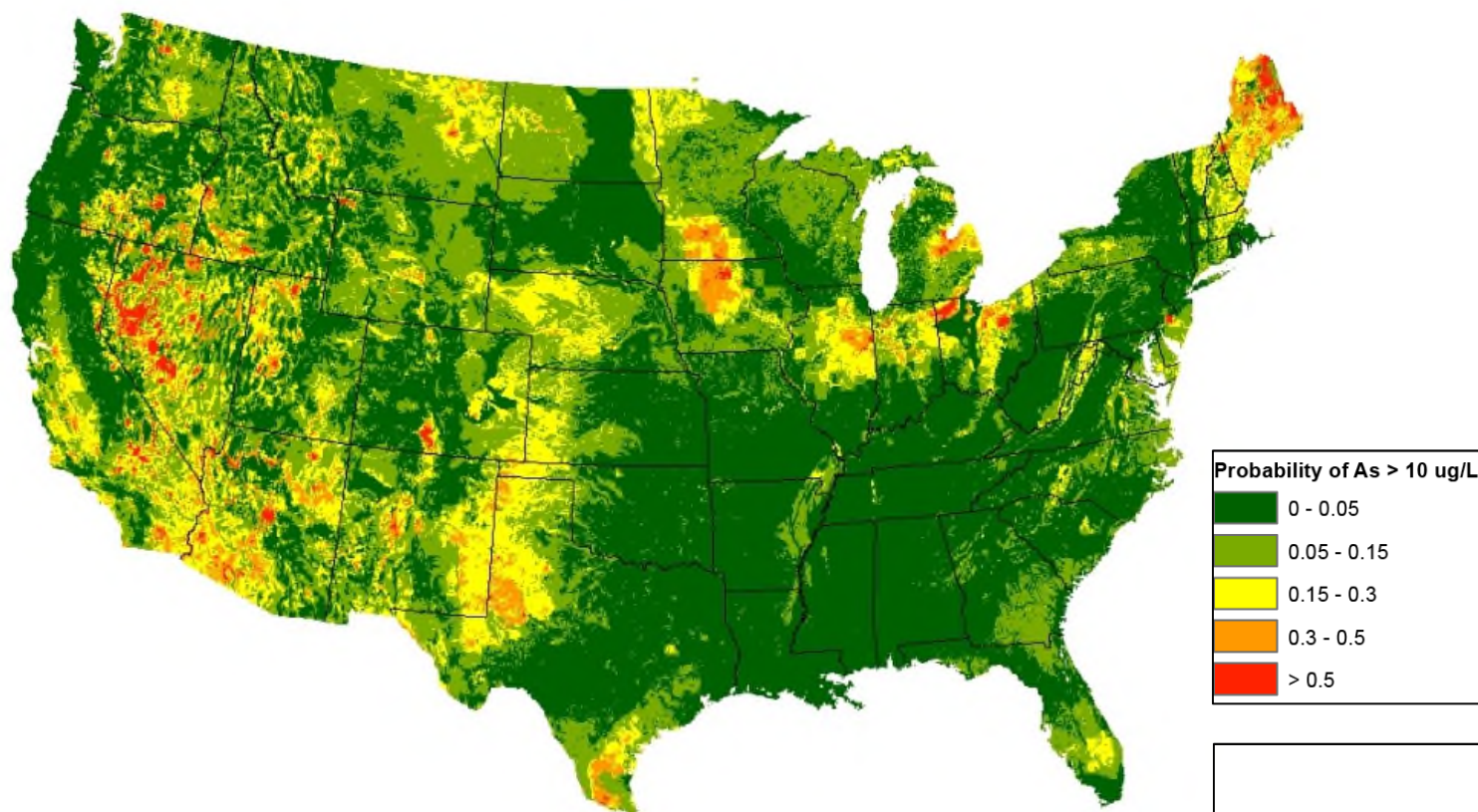


# A comparison of statistical modeling techniques to predict arsenic in domestic wells in the CONUS

Melissa Lombard, Bernard T. Nolan, Mathew Gribble, Maria Argos, Joseph Ayotte

## Estimating the High-Arsenic Domestic-Well Population in the Conterminous United States

Joseph D. Ayotte,<sup>\*,†,§</sup> Laura Medalie,<sup>‡</sup> Sharon L. Qi,<sup>§</sup> Lorraine C. Backer,<sup>||</sup> and Bernard T. Nolan<sup>⊥</sup>



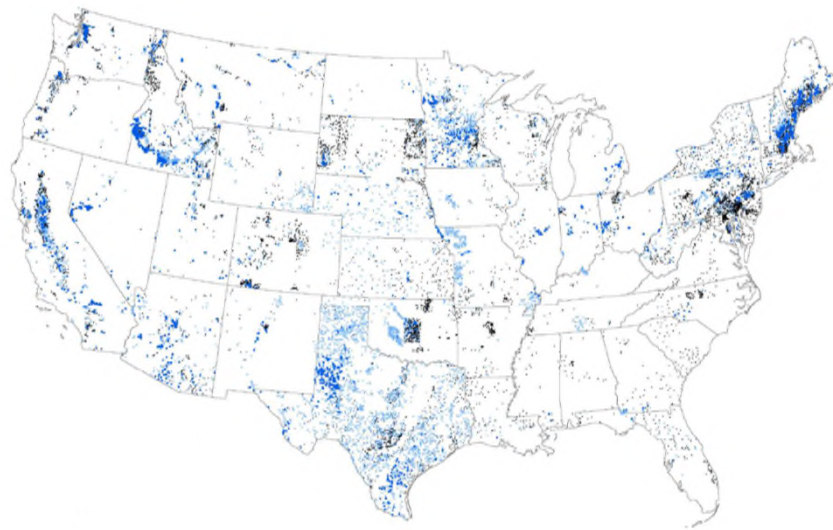
- Logistic Regression model
- Model response term is the probability of arsenic > 10µg/L

$$P(y|x) = \frac{e^{\beta_0 + \beta_1 x_1 + \dots + \beta_n x_n}}{1 + e^{\beta_0 + \beta_1 x_1 + \dots + \beta_n x_n}}$$



# Existing logistic regression model

Arsenic concentrations  
from 20,450 domestic wells



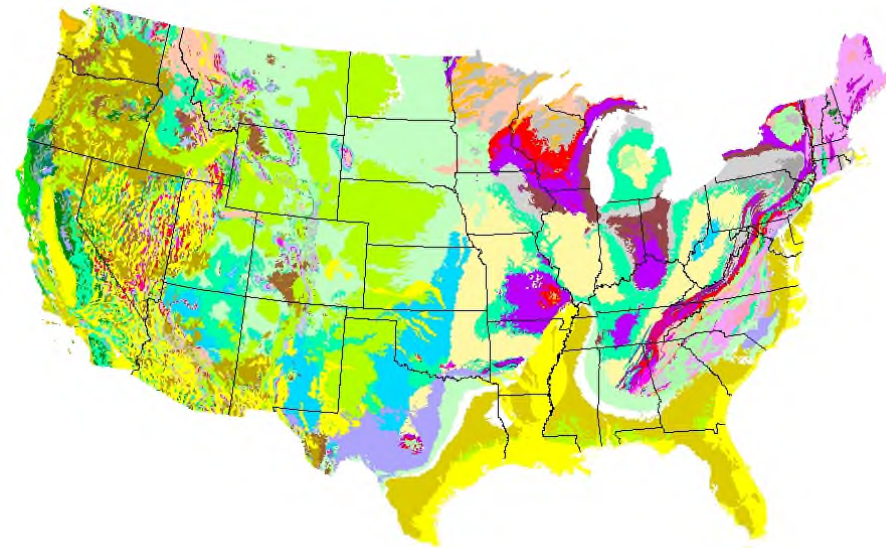
Black = As < 1 µg/L

Light blue =  $1 \leq \text{As} < 10 \text{ µg/L}$

Dark blue = As > 10 µg/L

*From Ayotte et al. 2017, ES&T*

42 predictor variables

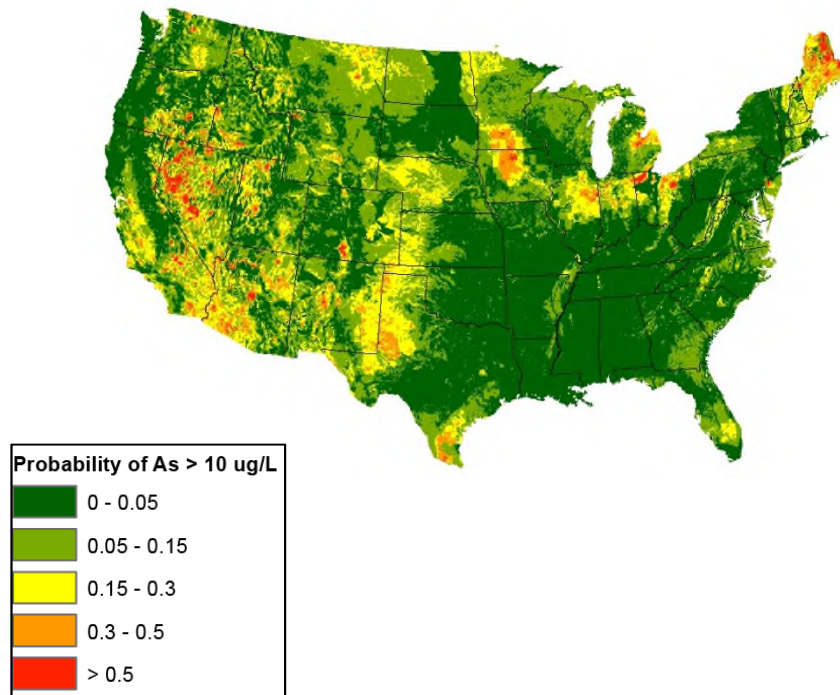


Generalized geology of the U.S.

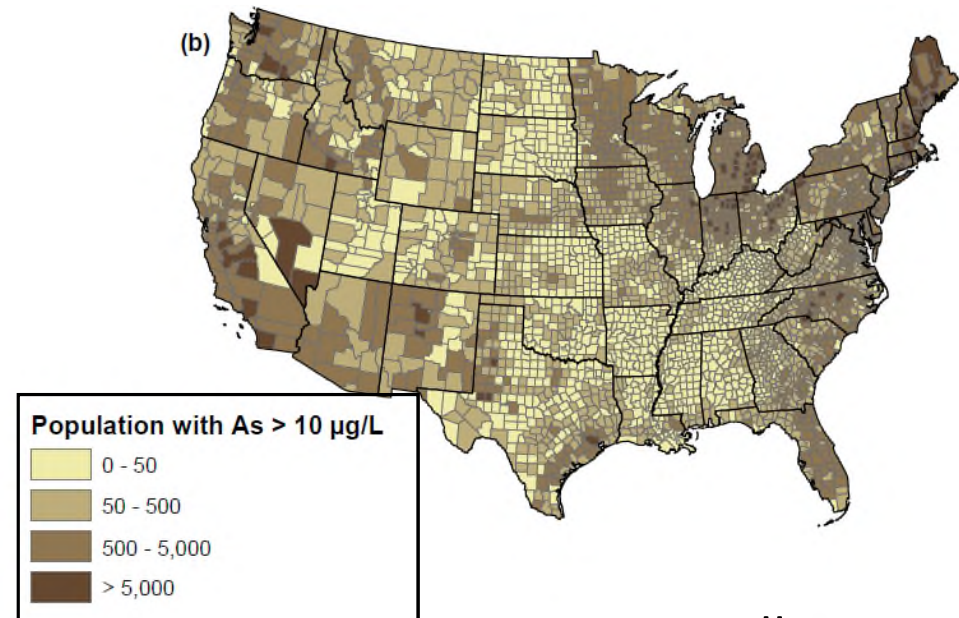
*After King and Beikman, 1974*

# Existing arsenic model and exposure estimate

Arsenic model output



Arsenic exposure estimate



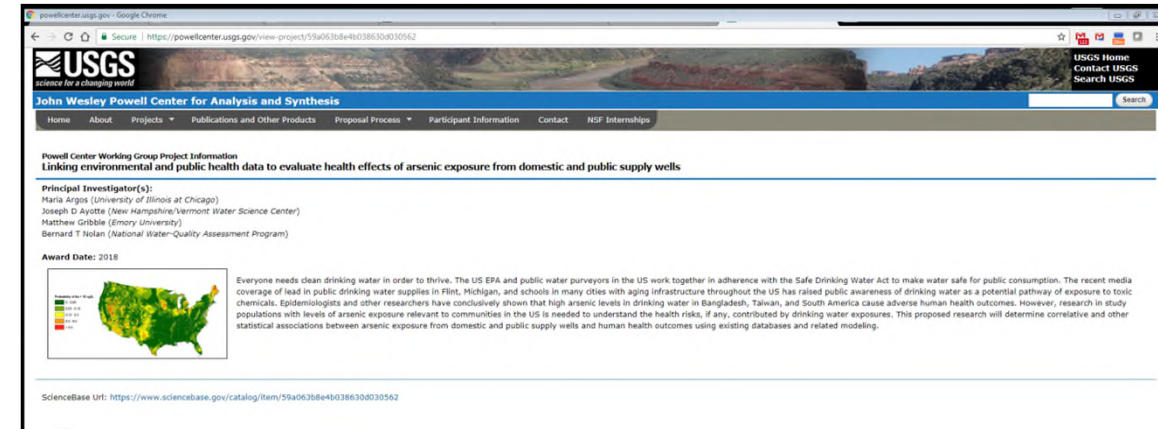
2.1 million people

*From Ayotte et al. 2017, ES&T*

# USGS Powell Center study

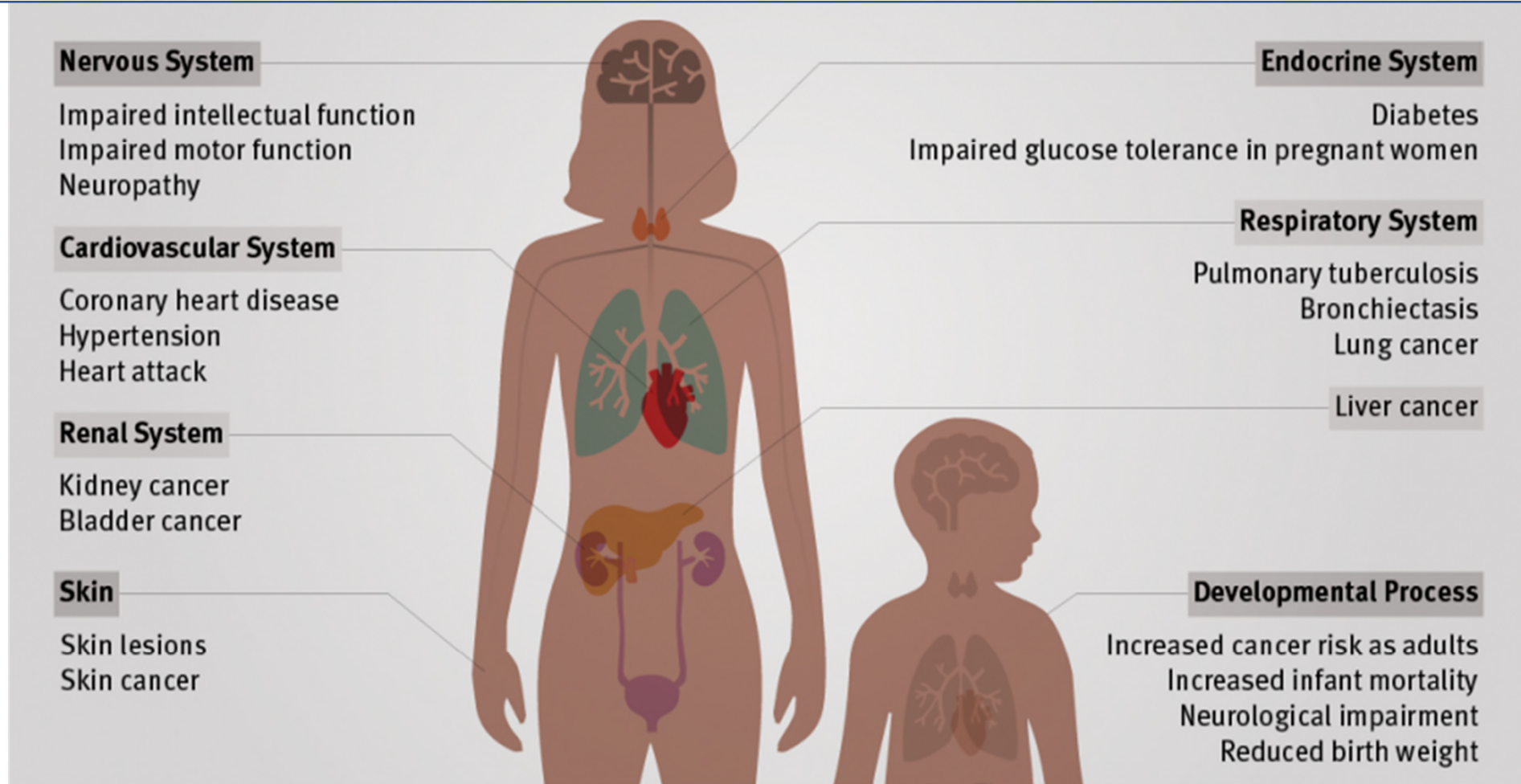
## Linking environmental and public health data to evaluate health effects of arsenic exposure from domestic supply wells

- Make a new national arsenic model
  - Use machine learning methods
  - Update model variables
- Results will be used in epidemiology models to evaluate relationships between human health outcomes and arsenic in domestic wells.



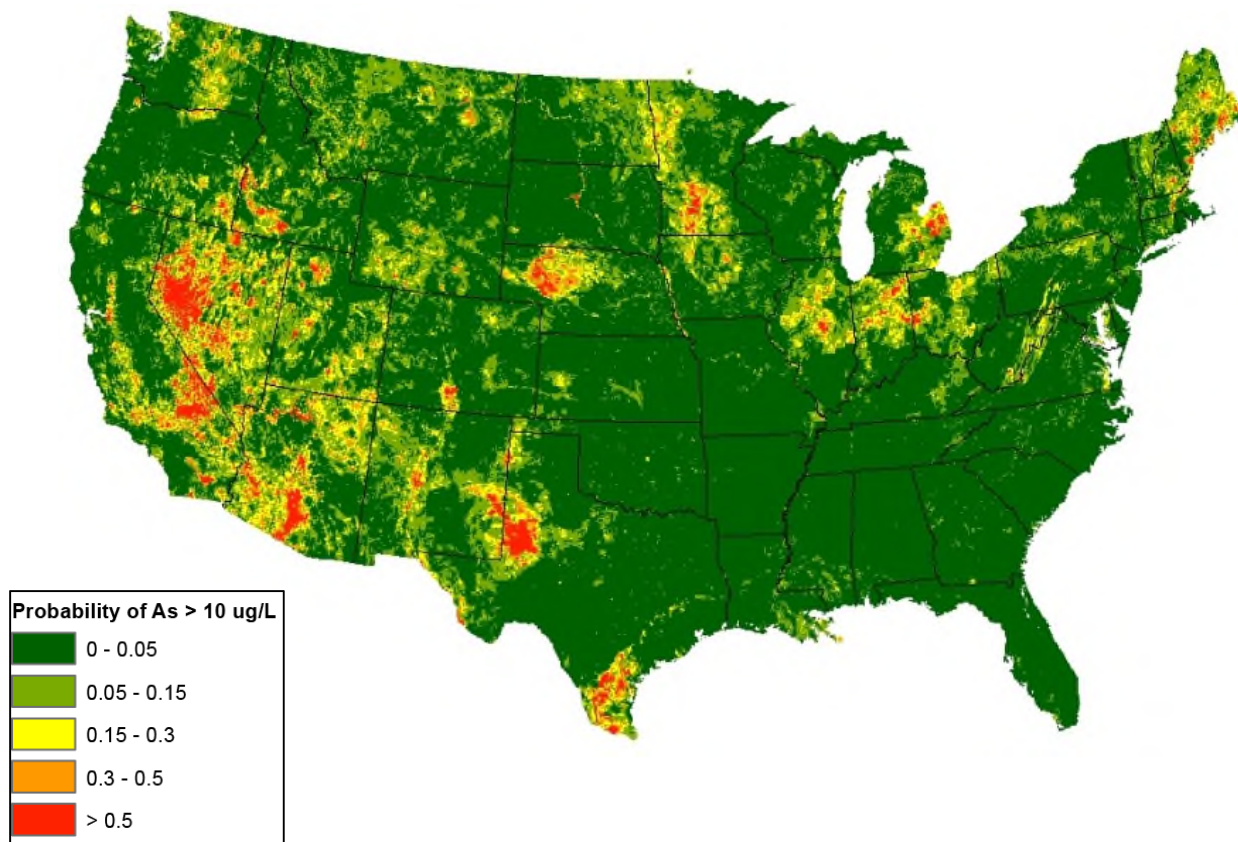


# Arsenic's effects on the human body



<https://www.hrw.org/report/2016/04/06/nepotism-and-neglect/failing-response-arsenic-drinking-water-bangladeshs-rural#page>

# Boosted Regression Tree Model

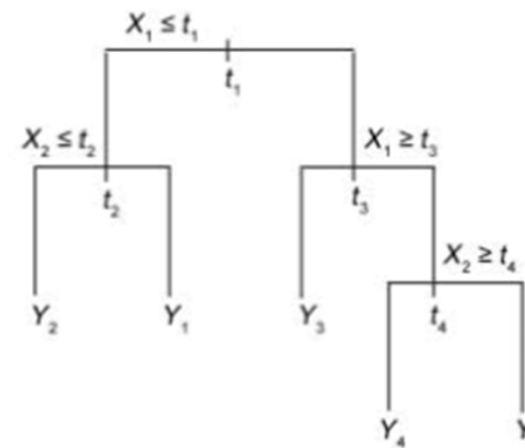


BRT Model

Number of trees = 4500

Interaction depth = 14

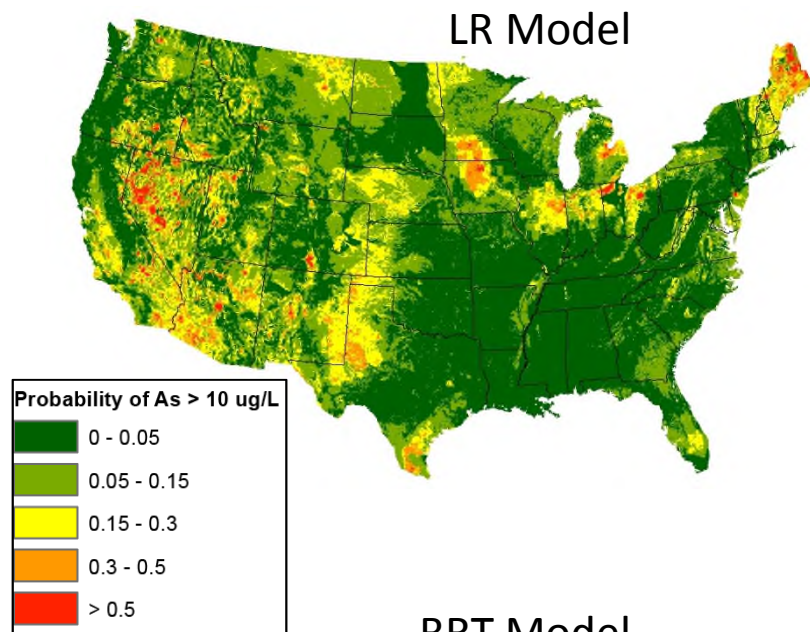
Learning rate = 0.01



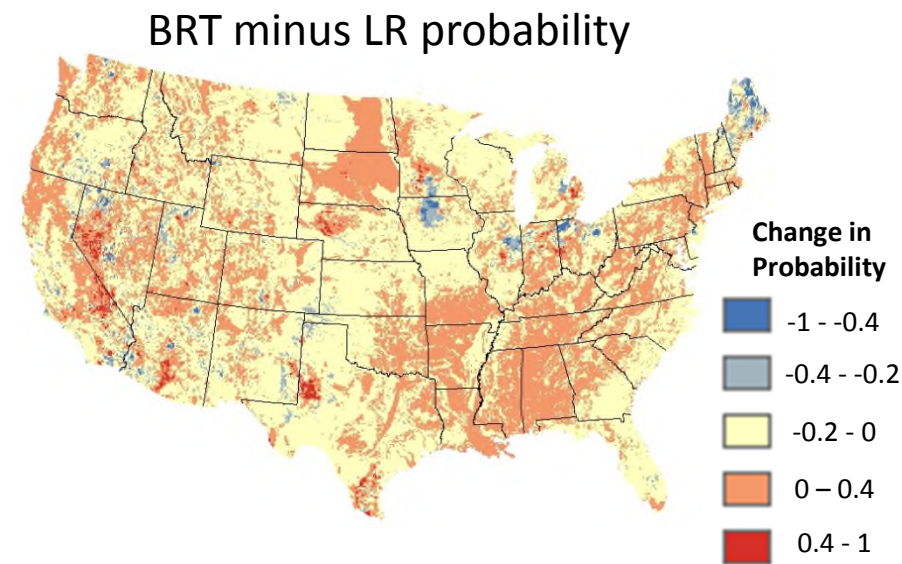
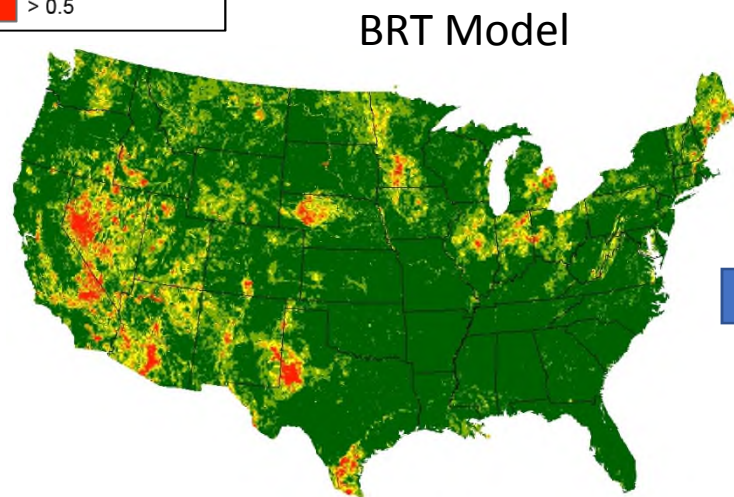
*From Elith et al., 2008*



# LR and BRT Model Comparison



- Sharper delineation between high and low arsenic probabilities with BRT model





# Model Comparison Predictive Performance

	Training Data			Hold-out Data		
	Accuracy	Sensitivity	Specificity	Accuracy	Sensitivity	Specificity
LR	89.9%	12.7%	99.3%	90.1%	13.9%	99.0%
BRT	95.6%	64.5%	99.3%	92.1%	43.7%	97.8%

Sensitivity = events ( $\text{As} > 10\mu\text{g/L}$ )

Specificity = non-events ( $\text{As} < 10\mu\text{g/L}$ )

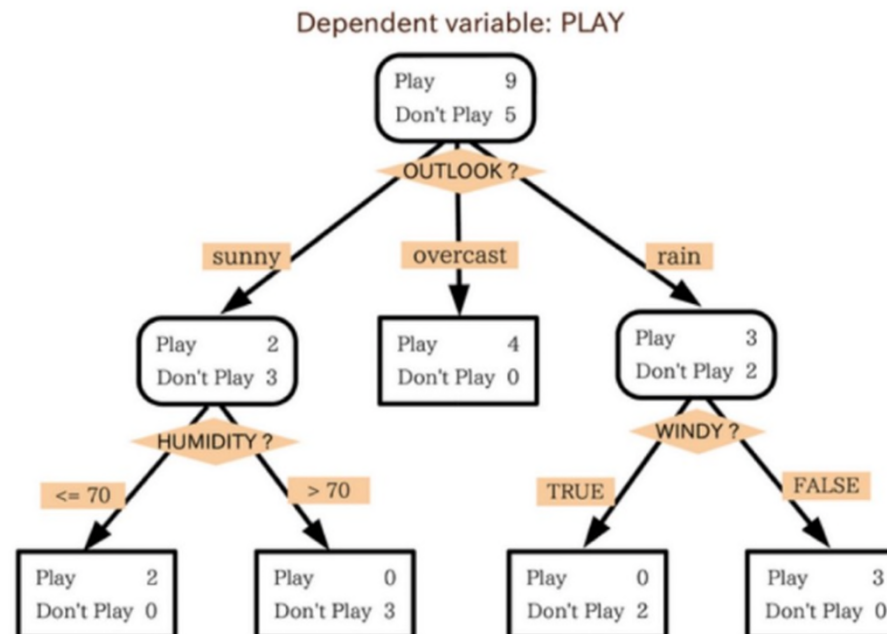
# Random Forest Classification Model

- Ensemble tree based machine learning method
- Model response term is a classification (category)
  - Arsenic  $\leq 10 \mu\text{g/L}$
  - Arsenic  $> 10 \mu\text{g/L}$

RFC Model

Number of trees = 500

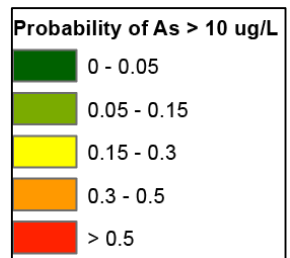
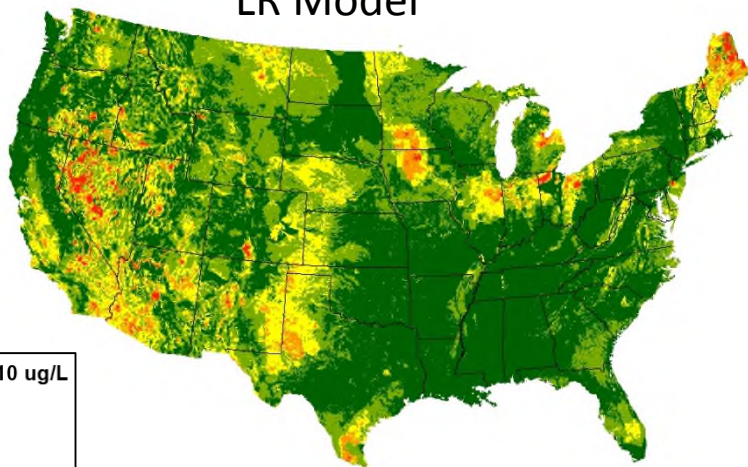
mtry = 38



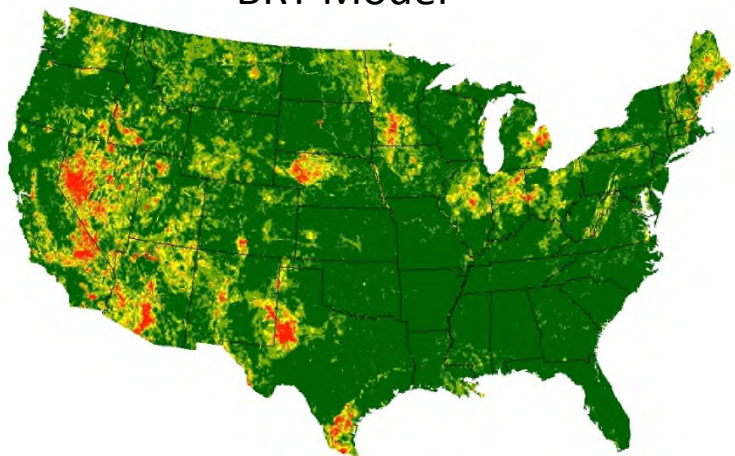
From medium.com

# LR, BRT, & RFC Model Comparison

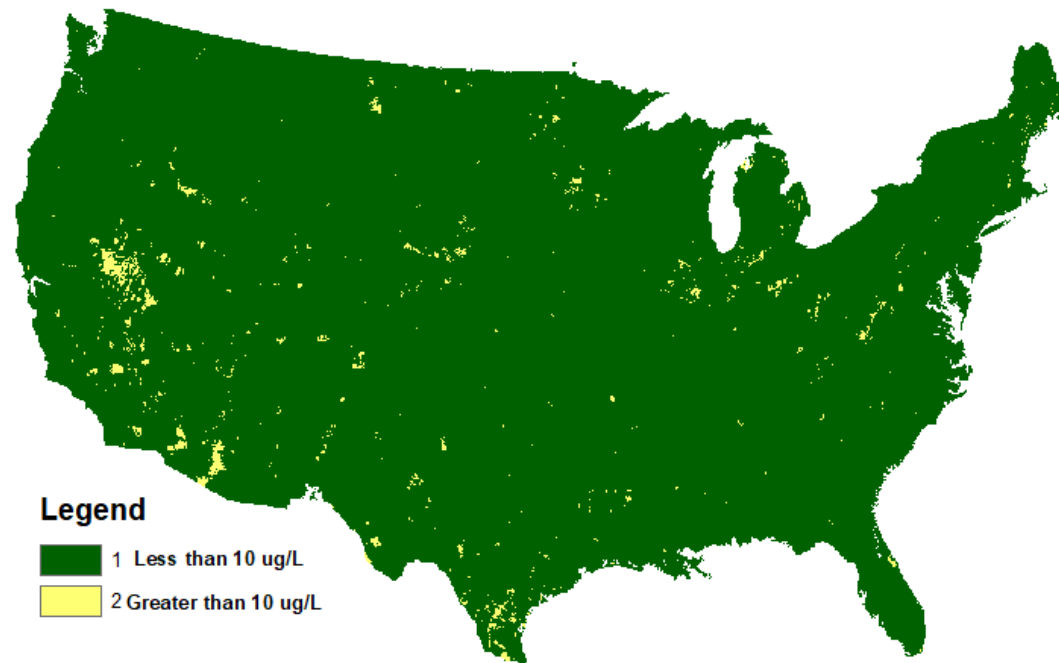
LR Model



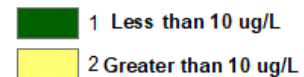
BRT Model



RFC Model



Legend





# Model Comparison Predictive Performance

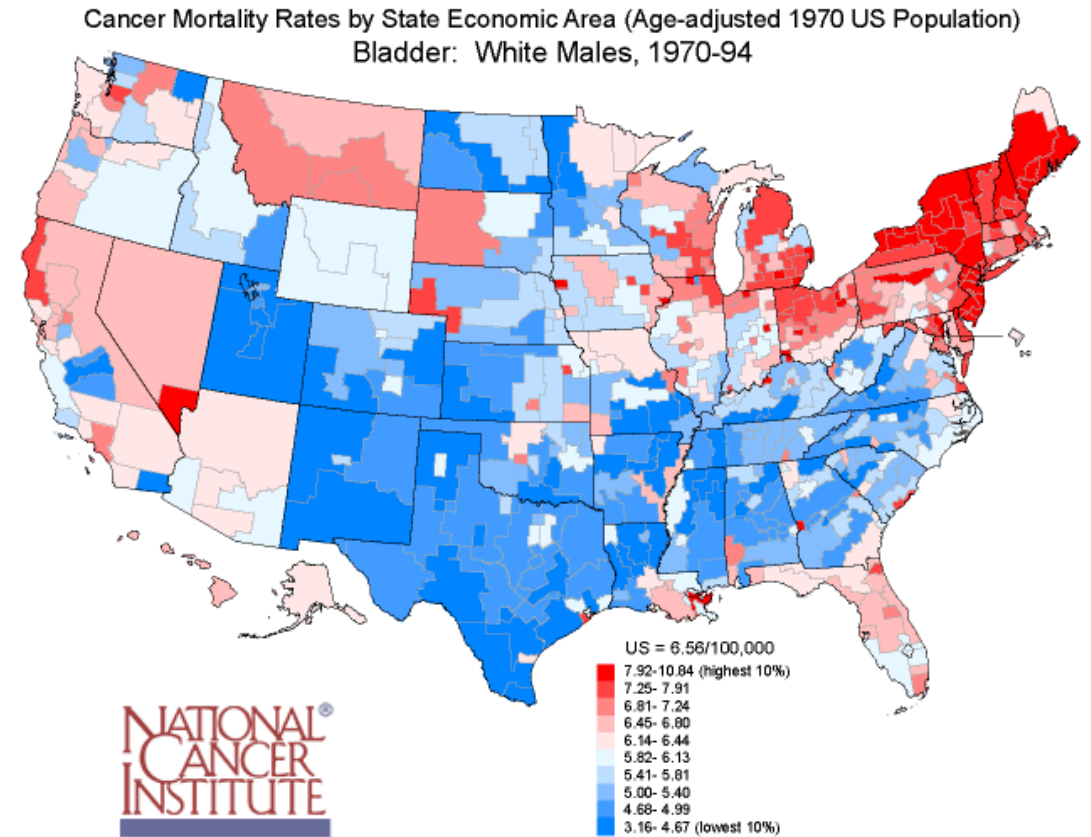
	Training Data			Hold-out Data		
	Accuracy	Sensitivity	Specificity	Accuracy	Sensitivity	Specificity
<b>LR</b>	89.9%	12.7%	99.3%	90.1%	13.9%	99.0%
<b>BRT</b>	95.6%	64.5%	99.3%	92.1%	43.7%	97.8%
<b>RFC 2C</b>	99.9%	100%	99.3%	91.0%	38.4%	97.1%

Sensitivity = events ( $\text{As} > 10\mu\text{g/L}$ )  
Classification  $\text{As} > 10\mu\text{g/L}$

Specificity = non-events ( $\text{As} < 10\mu\text{g/L}$ )  
Classification  $\text{As} < 10\mu\text{g/L}$

# Next steps

- Developing a RFC model with 4 concentration categories
  - $\leq 5 \mu\text{g/L}$
  - $5 - 10 \mu\text{g/L}$
  - $10 - 50 \mu\text{g/L}$
  - $> 50 \mu\text{g/L}$
- Arsenic model results will be used in epidemiology models
  - Low birth weight
  - Cancers
  - Diabetes
  - Cardiovascular disease



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# Questions ?